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STRONG LAW FOR MIXING SEQUENCE*

Xiru Chen and Yuehua Wu

Center for Multivariate Analysis University of Pittsburgh

Technical Report No. 87-47

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Center for Multivariate Analysis Fifth Floor Thackeray Hall University of Pittsburgh Pittsburgh, PA 15260



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ABSTRACT

In this note we present some theorems on the strong law for the mixing sequence which is not necessarily stationary, and the mixing coefficient involving only a pair of variables in the sequence.

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Key words and phrases: mixing coefficient, stationary sequence, strong law of large numbers.

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1. INTRODUCTION

In this article we present some results concerning the strong law of a mixing sequence $\{X_n, n \ge 1\}$. We do not assume that $\{X_n\}$ is stationary, and we use mixing coefficients involving only a pair of variables X, Y (in that order): The Rosenblatt mixing coefficient

$$\alpha(X,Y) = \sup\{|P(X \in A, Y \in B) - P(X \in A)P(Y \in B)|: A \in B', B \in B'\}$$

and the Ibragimov mixing coefficient

$$\beta(X,Y) = \sup\{|P(Y \in B | X \in A) - P(Y \in B)|: A \in B', B \in B', P(X \in A) > 0\}$$
 where B' is the σ -field of all Borel sets in R'.

THEOREM 1. Suppose that $\{X_n, n \ge 1\}$ is a sequence of random variables, and for some p > 1 the following conditions are satisfied:

$$\bar{i}^{\circ}$$
. $\sup_{n} E |X_{n}|^{p} < \infty$. (1)

 2° . There exists $\epsilon > 0$ such that as $|i - j| \rightarrow \infty$,

$$\alpha(X_{j}, X_{j}) \leq \rho(|i-j|) = \begin{cases} 0(|i-j|^{-p/(2p-2)-\epsilon}), & 1 (2)$$

Then

$$\lim_{n\to\infty} (S_n - ES_n)/n = 0, \quad a.s.$$
 (3)

Here and in the sequel $S_n = \sum_{i=1}^n X_i$.

THEOREM 2. Suppose that $\{X_n, n \ge 1\}$ is a sequence of random variables, or and one of the following conditions are satisfied:

$$(I) \quad \sum_{n=1}^{\infty} var(X_n)/n^2 < \infty, \quad \sup_{n} E|X_n| < \infty,$$
 and
$$\beta(X_j, X_j) \leq \mu(|i-j|), \quad \sum_{n=0}^{\infty} \mu^{1/2}(n) < \infty;$$

$$(4) \quad \text{v codes}$$

$$\sum_{n=0}^{\text{otto}} \mu(|i-j|), \quad \text{otto}$$

$$\sum_{n=0}^{\infty} \mu^{1/2}(n) < \infty;$$

$$\sum_{n=0}$$

(II) $\sup_{n} var(X_n) < \infty$ and there exists $\varepsilon > 0$ such that

$$\sum_{i=1}^{n} \mu^{1/2}(i) = O(n/(\log n)^{1+\varepsilon}); \qquad (5)$$

(III) (4) holds, X_1 , X_2 , ... are identically distributed and $E|X_1| < \infty$ (the existence of variance is not assumed). Then (3) is true.

Remarks:

- 1. Part (I) of Theorem 2 can be compared with a result of Blum et al [1], who assumes that $\{X_n\}$ is a *-mixing sequence instead of (4). Note that this assumption does not follow from (4). We can easily construct a pairwise independent sequence which is not *-mixing.
- 2. Parts (I) and (II) of Theorem 2 can also be compared with some results (see Theorem 3.7.2 and Theorem 3.7.4 of Stout [5]) derived from Serfling [4]. The conditions of these results involve correlation coefficients between two variables in the sequence.
- 3. Part (III) of Theorem 2 extends Theorem 1 of Etemadi [2]. The assumption that $\{X_i\}$ is identically distributed can be somewhat relaxed, for example, it can be replaced by the condition that there exists a random variable Y such that $P(|X_n| \ge x) \le P(|Y| \ge x)$ for all $n \ge 1$ and $x \ge 0$. We also mention a related result of Blum et at [1] Theorem 1. They assume that $\{X_n\}$ is identically distributed, the distribution of X_1 has a moment generating function in the neighborhood of zero and that $\{X_n\}$ is *-mixing. Under these more stronger conditions they prove that $P(|S_n ES_n|/n \ge \epsilon)$ tends to zero exponentially.

PROOF OF THE THEOREMS

In deducing our results we shall borrow a trick from Etemadi [2]. The following well-known facts concerning $\alpha(X,Y)$ and $\beta(X,Y)$ will be used:

$$|\operatorname{cov}(X,Y)| \leq 10(\alpha(X,Y))^{\delta/(2+\delta)}(E|X|^{2+\delta}E|Y|^{2+\delta})^{1/(2+\delta)}, \quad \delta > 0$$
 (6)

$$|\operatorname{cov}(X,Y)| \leq 2(\beta(X,Y)\operatorname{var}(X)\operatorname{var}(Y))^{1/2}. \tag{7}$$

For a proof, see Ibragimov and Linnik [3]. Also it is trivially true that

$$\alpha\left(XI_{C}(X), YI_{D}(Y)\right) \leq \alpha(X,Y), \qquad \beta\left(XI_{C}(X), YI_{D}(Y)\right) \leq \beta(X,Y) \tag{8}$$

$$\alpha(X-a, Y-b) = \alpha(X,Y), \qquad \beta(X-a, Y-b) = \beta(X,Y), \qquad (9)$$

where C and D are Borel sets in R' and a, b are constants.

Proof of Theorem 1. In view of (9), by defining $X_n^+ = X_n I(X_n > 0)$, $X_n^- = -X_n I(X_n \le 0)$, $n \ge 1$, we can assume without loss of generality that $X_n \ge 0$, n > 1. Define

$$Y_{n} = (X_{n} - EX_{n})I(|X_{n} - EX_{n}| < n^{1/p+\epsilon_{1}}), \quad n \ge 1,$$

$$S_{n}^{*} = \sum_{i=1}^{n} (Y_{i} - EY_{i}), \quad (10)$$

where ϵ_1 > 0 is a constant to be chosen later.

From condition (1) we have $\sum_{n=1}^{\infty} P(X_n - EX_n \neq Y_n) < \infty$ and $\lim_{n \to \infty} EY_n = 0$. Therefore, (3) is equivalent to

$$\lim_{n\to\infty} S_n^*/n = 0, \quad a.s.$$
 (11)

Now fix $\alpha > 1$ and let $k_n = [\alpha^n]$. For positive integer m sufficiently large, there exists n such that $k_n \le m < k_{n+1}$, and $n \to \infty$ as $m \to \infty$. From (1) we have

$$\sup_{n} E|Y_{n}| \equiv C < \infty. \tag{12}$$

Here and in the sequel C is an unimportant constant which is allowed to change. Since $Y_n \geq 0$, it follows that

$$S_{m}^{*} - S_{k_{n}}^{*} \ge -(m - k_{n})C$$
, when $S_{m}^{*} < S_{k_{n}}^{*}$, $S_{m}^{*} - S_{k_{n}}^{*} \le S_{k_{n+1}}^{*} - S_{k_{n}}^{*} + (k_{n+1} - m)C$, when $S_{m}^{*} \ge S_{k_{n}}^{*}$.

Hence

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$$|S_{m}^{*}/m - S_{k_{n}}^{*}/k_{n}| \le \left|\frac{k_{n+1}}{k_{n}} \frac{S_{k_{n+1}}^{*}}{k_{n+1}} - \frac{S_{k_{n}}^{*}}{k_{n}}\right| + \frac{k_{n+1} - k_{n}}{k_{n}} C.$$
 (13)

From (13) it follows that if we have shown that

$$\lim_{n\to\infty} S_{k_n}^{*}/k_n = 0, \quad a.s.$$
 (14)

Then we would have

$$\limsup_{m\to\infty} |S_m^*/m| \leq (\alpha - 1)C$$
, a.s.

For any $\alpha > 1$, hence (11).

By Borel-Cantelli lemma, in order to prove (14), we have only to show that

$$\sum_{n=1}^{\infty} \operatorname{var}(S_{k_n}^{*})/k_n^2 < \infty.$$
 (15)

By (6), (8) and (9), we have for any $\delta > 0$:

$$Var(S_{k_{n}}^{*}) = \sum_{k,j=1}^{k_{n}} cov(Y_{i}, Y_{j})$$

$$\leq C \sum_{i,j=1}^{k_{n}} (\alpha(X_{i}, X_{j}))^{\delta/(2+\delta)} (E|Y_{i}|^{2+\delta} E(Y_{j}|^{2+\delta})^{1/(2+\delta)}.$$
 (16)

From (1) it follows that

$$E|Y_n|^{2+\delta} \le Cn$$
 $(2+\delta-p)(1/p+\epsilon_1)$ $n = 1,2,...$ (17)

First consider the case p > 2. From (2), (16) and (17) we obtain

$$var(S_{k_{n}}^{*}) \leq C \sum_{i,j=1}^{k_{n}} (\alpha(X_{i},X_{j}))^{\delta/(2+\delta)} (ij)^{(2+\delta-p)(1/p+\epsilon_{1})/(2+\delta)}$$

$$\leq C \sum_{i,j=1}^{k_{n}} (\alpha(X_{i},X_{j}))^{\delta/(2+\delta)} i^{2(2+\delta-p)(1/p+\epsilon_{1})/(2+\delta)}$$

$$\leq C \sum_{i,j=1}^{k_{n}} i^{-(2/p+\epsilon)\delta/(2+\delta)} \sum_{i=1}^{k_{n}} i^{2(2+\delta-p)(1/p+\epsilon_{1})/(2+\delta)}.$$
(18)

Noticing 2/p < 1, we can assume that $2/p + \epsilon < 1$. Hence from (18) we have

$$\operatorname{var}(S_{k_n}^{\star}) \leq \operatorname{Ck}_n^{-(2/p+\epsilon)\delta/(2+\delta)} + 2(2+\delta-p)(1/p+\epsilon_1)/(2+\delta) + 2 \tag{19}$$

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This inequality holds for any $\delta > 0$. Now we choose $\epsilon_1 \in (0, \epsilon/2)$, then

$$\lim_{\delta\to\infty} \{-(2/p+\epsilon)\delta/(2+\delta) + 2(2+\delta-p)(1/p+\epsilon_1)/(2+\delta)\} = -\epsilon + 2\epsilon_1 \equiv n < 0.$$

Therefore, choosing δ sufficiently large, from (19) we obtain $\text{var}(S_{k_n}^{\star}) \leq Ck_n^{2-\eta}$. Hence (15) is true in view of $\sum_{n=1}^{\infty} k_n^{-\eta} < \infty$.

Next assume that p=2. Again, choose $\varepsilon_1\in (0,\varepsilon/2)$. Choose $\delta>0$ sufficiently small, such that $(1+\varepsilon)\delta/(2+\delta)<1$. We still have (19), with p=2. Since

$$-(1+\varepsilon)\delta/(2+\delta) + 2\delta(1/2+\varepsilon_1)/(2+\delta) = -(\varepsilon - 2\varepsilon_1)\delta/(2+\delta) < 0,$$

(15) holds again.

Finally, consider the case 1 . In this case we have, instead of (18),

$$var(S_{k_{n}}^{*}) \leq C \sum_{i=1}^{k_{n}} i^{-(p/(2p-2)+\epsilon)} \delta/(2+\delta) \sum_{i=1}^{k_{n}} i^{2(2+\delta-p)(1/p+\epsilon_{1})/(2+\delta)}.$$
 (20)

Write $\delta_0 = 2(p/(2p-2)-1+\epsilon)^{-1}$. Since $1 , we have <math>\delta_0 > 0$. Choose $\epsilon_1 > 0$ sufficiently small, such that

$$0 < \delta < \delta_0 \implies 2(2+\delta-p)(1/p+\epsilon_1)/(2+\delta) \le 1 - \eta$$

where $\eta > 0$ does not depend on δ , as long as $0 < \delta < \delta_0$. Because $(p/(2p-2)+\varepsilon)\delta/(2+\delta) < 1 \text{ for } 0 < \delta < \delta_0 \text{ and } (p/(2p-2)+\varepsilon)\delta_0/(2+\delta_0) = 1,$ one can find δ ε $(0,\delta_0)$, such that

$$1 - \eta/2 < (p/(2p-2) + \varepsilon)\delta/(2+\delta) < 1.$$

For this δ we have, by (20),

many accounts because basistees besides

$$var(S_{k_n}^*) \le Ck_n^{-(1-\eta/2)+1+(1-\eta)+1} \le Ck_n^{-\eta/2}.$$

So we obtain (15) again. Theorem 1 is proved.

Proof of Theorem 2. Part (I): Again we can assume $X_n \ge 0$. Write $Y_n = X_n - EX_n$ and $S_n^* = \sum_{i=1}^n Y_i$. From sup $E|X_n| < \infty$ we have sup $E|Y_n| < \infty$. Using the same argument employed in proving Theorem 1, we reduce the proof of (11) to that of (15). From (4), (7) and (9),

$$\sum_{n=1}^{\infty} var(S_{k_{n}}^{*})/k_{n}^{2} = \sum_{n=1}^{\infty} k_{n}^{-2} \sum_{j=1}^{k_{n}} cov(Y_{j}, Y_{j})$$

$$\leq C \sum_{n=1}^{\infty} k_{n}^{-2} \sum_{j=1}^{k_{n}} (\mu(|i-j|)var(X_{j})var(X_{j}))^{1/2}$$

$$\leq C \sum_{n=1}^{\infty} k_{n}^{-2} \sum_{j=0}^{k_{n}} \mu^{1/2}(i) \sum_{j=1}^{\infty} var(X_{j})$$

$$\leq C \sum_{n=1}^{\infty} k_{n}^{-2} \sum_{j=1}^{k_{n}} var(X_{j})$$

$$\leq C \sum_{n=1}^{\infty} var(X_{n})/n^{2} < \infty.$$
(21)

Part (II) is proved in much the same way as Part (I), only that we replace Ck_n for $\sum_{i=1}^k var(X_i)$ and $Ck_n/(\log n)^{1+\epsilon}$ for $\sum_{i=1}^k u^{1/2}(i)$ in (21) to obtain (22). Part (III) is proved by truncating X_n at n and combining the reasoning above and that of Etemadi [2].

AN EXAMPLE

Consider the autoregression model

$$X_n = a_1 X_{n-1} + ... + a_m X_{n-m} + e_n, \quad n = 0, \pm 1, \pm 2, ...$$
 (23)

We want to show that under certain conditions it is true that

$$\lim_{n\to\infty} \sum_{i=1}^{n} X_i/n = 0, \quad a.s.$$
 (24)

for any solution of (23). Suppose that the following conditions are satisfied:

1. $\{e_n, n = 0,\pm 1,...\}$ is a sequence of independent real random variables, and

$$Ee_n = 0, n = 0,\pm 1,..., \sup_{-\infty < n < \infty} E|e_n|^p = C < \infty \text{ for some p > 1.}$$
 (25)

where, as before, C is an unimportant constant which is allowed to change.

2. e_n has a density f_n satisfying the Lipschitz condition over R':

$$|f_n(x) - f_n(y)| \le C|x - y|, \quad n = 0, \pm 1, \pm 2, \dots$$
 (26)

where C does not depend on n.

3. a_1 , a_2 , ..., a_m are real constants, and the equation $1 - a_1 z - ... - a_m z^m = 0$ has all its root outside the unit circle.

Under the condition 1 and 3, the general real solution of (23) has

the form

$$X_{n} = \sum_{t=0}^{\infty} b_{t} e_{n-t} + \sum_{j=0}^{J} \rho_{j}^{n} \sum_{\ell=0}^{m_{j}-1} n^{\ell} (\xi_{j} \cos n\omega_{j} + \eta_{j} \sin n\omega_{j}) = \tilde{X}_{n} + X_{n}^{*}$$
 (27)

where $b_0 = 1$, b_2 , b_3 , ... are real constants such that

$$|b_t| \le CH^t$$
, $t = 0,1,2,...$ for some $H \in (0,1)$. (28)

 ho_j and ω_j , j = 1,...,J, are real constants, $0 < \rho_j < 1$, j = 1,...,J, m_l + ... + m_j = m, and ξ_j , η_j , ℓ = 1,..., m_j , j = 1,...,J, are arbitrary random variables. From (25), (27) and (28) it follows that

$$E\tilde{X}_{n} = 0, \quad n = 0, 1, 2, ..., \quad \sup_{-\infty < n < \infty} E |\tilde{X}_{n}|^{p} = C < \infty.$$
 (29)

Let n, N be positive integers, n < N. Define

$$Y_{nN} = \sum_{t=0}^{N-n-1} b_t e_{N-t}, \qquad Z_{nN} = \sum_{t=N-n}^{\infty} b_t e_{N-t}.$$

Since $b_0 = 1$, from (26) it follows that the density g_{nN} of Y_{nN} obeys Lipschitz's condition with the same constant C as in (26). Also

$$\sup\{E|Y_{nN}|^p: 1 \le n < N < \infty\} = C < \infty.$$
 (30)

Now let q_1 be a positive constant, $q_2 = 2q_1$. Define the event

$$D_{nN} = \{ |Z_{nN}| \ge (N-n)^{-q_2} \}. \tag{31}$$

(25) entails $\sup_{-\infty < n < \infty} E|e_n| = C < \infty$. Hence

$$P(D_{nN}) \le C(N-n)^{q_2} \sum_{t=N-n}^{\infty} H^t \le C(N-n)^{q_2} H^{N-n}.$$
 (32)

Let G be a Borel set in R', h be a constant. G - h is defined as the

set $\{g-h: g\in H\}$. Write $\tilde{G}=G \cap \{u: |u| \leq (N-n)^{q_j}\}$, $G^*=G \setminus \tilde{G}$. If |h| < 1, we have

$$\begin{split} |P(Y_{nN} \in G) - P(Y_{nN} \in G - h)| \\ &\leq |P(Y_{nN} \in \widetilde{G}) - P(Y_{nN} \in \widetilde{G} - h)| + P(Y_{nN} \in G^*) + P(Y_{nN} \in G^* - h) \\ &\leq \int_{\widetilde{G}} |g_{nN}(u) - g_{nN}(u - h)| du + P(|Y_{nN}| > (N - n)^{q_1}) + P(|Y_{nN}| > (N - n)^{q_1} - 1) \\ &\leq C(N - n)^{q_1} h + C(N - n)^{-q_1} + C[(N - n)^{q_1} - 1]^{-1} \\ &\leq C(N - n)^{q_1} h + C(N - n)^{-q_1}. \end{split}$$

$$(33)$$

Now let A and B be two Borel sets in R'. We proceed to estimate $|P(\widetilde{X}_n \in A, \ \widetilde{X}_N \in B) - P(\widetilde{X}_n \in A)P(\widetilde{X}_N \in B)|.$ From (32), (33) and the independence of e_1 , e_2 , ..., we have

$$|P(\tilde{X}_{n} \in B | e_{n}, e_{n-1}, \dots) - P(Y_{nN} \in B)| = |P(Y_{nN} \in B - Z_{nN} | Z_{nN}) - P(Y_{nN} \in B)|$$

$$\leq C(N-n)^{-(q_{2}-q_{1})} + C(N-n)^{-q_{1}}$$

$$\leq C(N-n)^{-q_{1}}, \qquad (34)$$

when D_{nN} does not occur. But

$$\begin{split} |P(\tilde{X}_{N} \in B) - P(Y_{nN} \in B)| &= |P(Y_{nN} \in B - Z_{nN}) - P(Y_{nN} \in B)| \\ &= |P(D_{nN}^{C})P(Y_{nN} \in B - Z_{nN}) + P(D_{nN})P(Y_{nN} \in B - Z_{nN}|D_{nN}) \\ &- P(Y_{nN} \in B)| \\ &\leq P(D_{nN}) + |P(Y_{nN} \in B - Z_{nN}|D_{nN}^{C}) - P(Y_{nN} \in B)| + P(D_{nN}) \\ &\leq 2P(D_{nN}) + C(N-n)^{-q_{1}} \leq C(N-n)^{q_{2}}H^{N-n} + C(N-n)^{-q_{1}} \\ &\leq C(N-n)^{-q_{1}}. \end{split}$$

$$(35)$$

From (34) and (35) we get

$$|P(\tilde{X}_{N} \in B | e_{n}, e_{n-1}, \dots) - P(\tilde{X}_{N} \in B)| \leq C(N-n)^{-q_{1}}$$

when D_{nN} does not occur. If $P(\tilde{X}_n \in B) \ge C(N-n)^{-q_1}$, then from (33) and (35) we obtain

$$P(\tilde{X}_{n} \in A, \tilde{X}_{N} \in B) \ge [P(\tilde{X}_{N} \in B) - C(N-n)^{-q_1}][P(\tilde{X}_{n} \in A) - C(N-n)^{q_2}H^{N-n}]. \tag{36}$$

Also

$$P(\tilde{X}_{n} \in A, X_{N} \in B) \leq [P(\tilde{X}_{N} \in B) + C(N-n)^{-q_{1}}][P(\tilde{X}_{n} \in A) + C(N-n)^{q_{2}}][P(\tilde{X}_{n} \in A) + C(N-n)^{q_{2}}][P(\tilde{X}_$$

From (36) and (37) we have

$$|P(\tilde{X}_{n} \in A, \tilde{X}_{N} \in B) - P(\tilde{X}_{n} \in A)P(\tilde{X}_{N} \in B)| \leq C(N-n)^{-q_{1}} + C(N-n)^{q_{2}}H^{N-n} + C(N-n)^{q_{1}}H^{N-n} \leq C(N-n)^{-q_{1}}, \quad (38)$$

where C does not depend on A, B. (38) is proved when $P(\tilde{X}_n \in B) \ge C(N-n)^{-q_1}$. If $P(\tilde{X}_N \in B) < C(N-n)^{-q_1}$, (38) is trivially true. Therefore we get

$$\alpha(\tilde{X}_{n},\tilde{X}_{N}) \leq C(N-n)^{-q_{1}}. \tag{39}$$

Now choose $q_1 = p/(2p-2) + 2$. From (39) we see that the condition (2) is satisfied. This, together with (29), gives, by Theorem 1,

$$\lim_{n\to\infty} \sum_{j=1}^{n} \tilde{X}_{j}/n = 0, \quad a.s.$$
 (40)

From the expression of X_n^* , it is readily seen that

$$\lim_{n\to\infty} \sum_{j=1}^{n} X_{j}^{*}/n = 0, \quad a.s.$$
 (41)

From (27), (40) and (41), we obtain (24).

The conclusion (40) does not follow from the ergodic theorem of stationary process, since $\{e_n\}$ is not assumed to be identically distributed, so $\{X_n\}$ may not be a strictly stationary process.

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